

A COMPUTER ROUTINE FOR THE ESTIMATION OF VARIANCE COMPONENTS IN THE  
GENERAL MIXED MODEL BY THE RESTRICTED MAXIMUM LIKELIHOOD (REML) PROCEDURE

by

BU-543-M

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Abstract

This documentation supports a program which computes estimates of fixed effects and variance components in the general mixed model by the REML procedure described in Corbeil and Searle (1974). The program language is Fortran IV for the IBM systems 360 and 370 series. A listing of the source code is appended to the documentation.

Introduction

The theory of restricted maximum likelihood (REML) estimation of variance components in the general mixed model and the algorithm which forms the basis of the computer routine is fully described in Corbeil and Searle (1974). The model for  $\underline{y}$ , a vector of  $N$  observations, is

$$\underline{y} = \underline{X}\underline{\mu} + \underline{U}_1 \underline{b}_1 + \cdots + \underline{U}_c \underline{b}_c + \underline{e}$$

and

$$\underline{y} \sim N\left(\underline{X}\underline{\mu}, \sum_{i=1}^c \sigma_i^2 \underline{U}_i \underline{U}_i' + \underline{I}_N \sigma^2\right),$$

where

$\underline{\mu}$  is a vector of  $k$  unknown constants, the fixed effects of the model,

$\underline{X}$  is an  $N \times k$  incidence matrix, of full column rank, corresponding to

$\underline{\mu}$  and with  $k < N$ ,

$\underline{U}_i$  is an  $N \times m_i$  design matrix associated with the  $i^{\text{th}}$  random factor, with  $\sum_{i=1}^c m_i + k < N$ ,

$\underline{b}_i$  is a vector of  $m_i$  random variables which are i.i.d.  $N(0, \sigma_i^2)$ , with the  $\underline{b}_i$ 's being mutually independent,

and

$\underline{e}$  is a vector of  $N$  random variables which are i.i.d.  $N(0, \sigma^2)$  and independent of the  $\underline{b}_i$ 's.

As part of the necessary information to be provided to the program is the  $\underline{y}$  vector of  $N$  observations, the matrices  $\underline{U}_1, \dots, \underline{U}_c, \underline{X}$  and the constants  $N, k, c$ , and all  $m_i$ 's. The program will compute by an iterative procedure the estimates of  $c$  variance components in the form of  $\gamma_i = \sigma_i^2/\sigma^2$ ,  $\sigma^2$ , an objective function (which is being minimized) and the  $c$  first order partials. These are printed at each iteration. Upon the simultaneous convergence of the  $c$   $\gamma$ -components, estimates of  $\underline{\mu}$  are computed. These are also printed along with the  $\gamma_i$ 's, the  $\sigma_i^2$ 's, and their estimated variances and covariances.

#### Description of input card deck

The user card deck consists of a title card, 2 or more control cards containing the constants of the model and several program control parameters, 2 format control cards, and the data.

##### 1. Title card (1<sup>st</sup> card)

This is a single card which contains a message in columns 1-80 that the user may wish to have appear as a heading to his output.

##### 2. Control cards

(a) The 2<sup>nd</sup> card contains the following constants in the format 7I5:

$N$  = the total number of observations

$K = k$ , the number of fixed effects in the model

$IC = c$ , the number of random effects in the model

$M = \sum_{i=1}^c m_i$ , the total number of levels of all random factors

$IT$  = the maximum number of iterations to be allowed should the stopping rule fail

$IOUT = \begin{cases} 0 \text{ or blank, if the input data } (U_1 \cdots U_{\sim c} Xy) \text{ is not to be listed} \\ \text{some integer value, if the input data is to be listed} \end{cases}$

$IPREC$  = the number of significant digits required of the estimation procedure, the default value is 10

- (b) The 3<sup>rd</sup> card and possibly one or more subsequent cards contain the  $c$   $m_i$  values of the model in the format 16I5, where  $m_i$  is the number of levels of the  $i^{th}$  random factor, i.e., the number of columns in  $U_i$ . The  $m_i$ 's should be listed in numerical sequence for  $i = 1, \dots, c$  to a maximum of 16 per card.

### 3. Format cards

There are two format cards which follow in sequence after the control cards described above:

- (a) The data which is taken to be in the form of an augmented matrix  $(U_1 \cdots U_{\sim c} Xy)$  is read one row at a time and a format card, e.g., (6F3.0,F5.1), is necessary for locating each of its elements.
- (b) Another format card, e.g. (1X,5F3.0,3X,F3.0,3X,F5.1), controls the listing of the augmented matrix if the IOUT option is  $> 0$ . If  $IOUT = 0$  or blank, a blank card or dummy format card is necessary to assure the proper sequence of cards is read by the program.

#### 4. Data cards

- (a) There are N sets of data cards to be read under the control of format card (3a) described above. Each set represents a row of the  $(U_{\sim 1} \cdots U_{\sim c} X_{\sim y})$  matrix.
- (b) The iterative procedure requires initial estimates of the  $\gamma_i$ 's. These are read in under the format 8F10.0. It is desirable to use the very best estimates available, but should none be available, setting each  $\gamma_i = 1.0$  is usually adequate. NOTE: If the initial estimates of the  $\gamma_i$ 's are such that the objective function increases for 2 successive iterations, processing is discontinued.

#### Program description

The program is written in 4 parts: a mainline which contains the dimension statement, a subroutine which carries out the bulk of the calculations, and two inversion subroutines, each being applied to a special task. Although a source listing is attached, the strategy which is prescribed by these subroutines is detailed in the appendix of Corbeil and Searle (1974).

The minimum dimensions to be specified in the mainline are:

MI (c + 1),

R (m + k + 1, m + k + 1), where  $m = \sum_{i=1}^c m_i$ ,

W (m + k + 1, m + k + 1),

D (c),

WM (f, f), where  $f = \max(m, k + 1)$ ,

WM2 (c, f),

DR (c),

and TITLE (20),

with c and k as previously defined.

```
//XF981      JOB XXXXXXXX,'810--R. R. CORBEIL  ',PRTY=6      NLVR=WRN
/*LIMITS REGION=100K,CLASS=C,CPUTIME=(,20)
// EXEC FORTGCLG
```

```
C*****RESTRICTED MAXIMUM LIKELIHOOD (REML) ESTIMATION OF VARIANCE
C*****COMPONENTS FOR THE MIXED MODFL (BU-538-M) BASED ON THE HEMMERLE
C*****AND HARTLEY ALGORITHM (TECHNOMETRICS 15:819-831, 1973). THE
C*****DOCUMENTATION FOR THIS PROGRAM IS (BU-543-M), A BIOMETRICS UNIT,
C*****CORNELL UNIVERSITY, ITHACA, N. Y. MIMEOGRAPH SERIES.
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```
      IMPLICIT REAL*8(A-H,O-Z)
      DIMENSION MI(2),R(10,10),W(10,10),D(1),WM(7,7),WM2(1,7),
1DR(1),TITLE(20)
120  READ(5,102,END=999) (TITLE(I),I=1,20)
102  FORMAT(20A4)
      WRITE(6,103) (TITLE(I),I=1,20)
103  FORMAT('1','***** REML: ',20A4,' *****')
      READ(5,110)N,K,IC,M,IT,IOUT,IPREC
110  FORMAT(7I5)
      WRITE(6,111)N,K,IC,M,IT
111  FORMAT('0N=',I5,' K=',I5,' C=',I5,' M=',I5,' IT=',I5)
      MI(1)=0
      K1=K+1
      K2=M+K1
      IC1=IC+1
      READ(5,112)(MI(I+1),I=1,IC)
112  FORMAT(16I5)
      WRITE(6,113)(I,MI(I+1),I=1,IC)
113  FORMAT(16I5)
      IF(M.GT.K1)GO TO 100
      MAXMK1=K1
      GO TO 101
100  MAXMK1=M
101  CALL PMAIN(MI,R,W,D,DR,WM,WM2,N,K,IC,M,IT,K1,K2,IC1,MAXMK1,
+IOUT,IPREC)
      GO TO 120
999  STOP
      END
```

```
      SUBROUTINE PMAIN(MI,R,W,D,DR,WM,WM2,N,K,IC,M,IT,K1,K2,IC1,MAXMK1,
+IOUT,IPREC)
      IMPLICIT REAL*8(A-H,O-Z)
      DIMENSION MI(IC1),R(K2,K2),W(K2,K2),D(IC),DR(IC),
1WM(MAXMK1,MAXMK1),WM2(IC,MAXMK1)
      DIMENSION FMT1(18),FMT2(18)
      READ(5,200)FMT1
      READ(5,200)FMT2
200  FORMAT(18A4)
      TOL=.5D-10
      EPS=.001D0
      PREC=1.0D1**(-IPREC)
      IF(IPREC.EQ.0) PREC=1.0D-10
      IT1=0
      IEND=0
      TOBF=0.0D0
      M1=M+1
      K4=K2
      C      INPUT (V X Y) AND FORM R
      DO 1 I=1,K2
      DO 1 J=I,K2
1      R(I,J)=0.
      WRITE(6,202) IOUT
202  FORMAT('00OUTPUT DATA OPTION = ',I5)
      DO 2 L=1,N
      READ (5,FMT1) (W(1,I),I=1,K2)
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```

      IF (IOUT.GT.0) WRITE(6,FMT2) (W(1,I),I=1,K2)
      DO 2 I=1,K2
      DO 2 J=I,K2
2      R(I,J)=R(I,J)+W(1,I)*W(1,J)
C      REARRANGE SUBMATRICES OF R TO COINCIDE WITH ( V Y X )
      DO 59 I=1,K2
59      W(2,I)=R(I,K2)
      DO 64 I=1,M
      DO 64 J=1,K
64      R(I,K2-J+1)=R(I,K2-J)
      DO 68 I=1,K
      DO 68 J=1,I
68      R(K2-I+1,K2-J+1)=R(K2-I,K2-J)
      R(M1,M1)=W(2,K2)
      DO 69 I=1,M
69      R(I,M1)=W(2,I)
      DO 84 I=1,K
84      R(M1,M1+I)=W(2,M+I)
      DO 3 I=1,K2
      DO 3 J=I,K2
3      R(J,I)=R(I,J)
C      INPUT INITIAL VALUES OF D
29      READ(5,203)(D(I),I=1,IC)
203      FORMAT(8F10.0)
      DO 44 I=1,IC
44      DR(I)=DSQRT(D(I))
C      INVERT X'X PORTION OF R
      M2=M+2
      CALL INV2(P,K,0,M2,K2,DD,K2,M1)
C      TRANSFORM R ( M1 X M1 )
      DO 31 I=1,M
      DO 31 J=1,K
      W(I,J)=0.000
      DO 31 II=1,K
31      W(I,J)=W(I,J)+R(I,M1+II)*R(M1+II,M1+J)
      DO 32 I=1,M
      DO 32 J=1,M
      DO 32 II=1,K
32      R(I,J)=R(I,J)-W(I,II)*R(M1+II,J)
      DO 24 I=1,M
      DO 33 II=1,K
33      R(I,M1)=R(I,M1)-W(I,II)*R(M1+II,M1)
24      R(M1,I)=R(I,M1)
      YY=0.000
      DO 34 I=1,K
      W(1,I)=0.000
      DO 35 II=1,K
35      W(1,I)=W(1,I)+P(M1,M1+II)*R(M1+II,M1+I)
34      YY=YY+W(1,I)*R(M1+I,M1)
      R(M1,M1)=R(M1,M1)-YY
      K3=K2
      NK=N-K
      WRITE(6,209)
209      FORMAT('-'/'0','ITERATION NO.',8X,'OBJECTIVE FUNCTION',8X,'SIGMA S
      &QUARED',12X,'FIRST ORDER PARTIALS',16X,'GAMMA')
      GO TO 30
6      CONTINUE
C      COMPUTE SIGMA SQUARED
      SIGMA=W(K2,K2)
      SIGMA=SIGMA/NK
C      COMPUTE DEBJECTIVE FUNCTION (MODIFIED LIKELIHOOD)
      OBF=(N-K)*DLOG(SIGMA)+DET
      IF (OBF.LT.TORF) GO TO 401
      IEND=IEND+1
      GO TO 402
401      IEND=0

```

402 TOBF=UBF .  
C COMPUTE 1ST ORDER PARTIALS (NEGATIVES) A.3  
LL=1  
LU=0  
DO 11 I=1,IC  
LL=LL+MI(I)  
LU=LU+MI(I+1)  
WM(I,IC1)=0  
L1=1  
DO 12 L=LL,LU  
WM(I,IC1)=WM(I,IC1)+W(L,L)  
WM2(I,L1)=W(K2,L)  
12 L1=L1+1  
TEMP=0  
L1=1  
DO 14 L=LL,LU  
TEMP=TEMP+WM2(I,L1)\*WM2(I,L1)  
14 L1=L1+1  
TEMP=TEMP/(2.0D0\*SIGMA)  
WM(I,IC1)=(WM(I,IC1)/2.0D0)-TEMP  
11 WM(I,IC1)=DR(I)\*WM(I,IC1)  
C INTERMEDIATE OUTPUT  
WRITE(6,210) IT1,UBF,SIGMA,WM(1,IC1),D(1)  
210 FORMAT('U',10X,I2,11X,G18.10,3X,G18.10,6X,' 1 ',2X,G18.10,9X,  
& G18.10)  
IF (IC.EQ.1) GO TO 13  
DO 48 I=2,IC  
WRITE(6,211) I,WM(I,IC1),D(I)  
211 FORMAT(' ',70X,I3,3X,G18.10,9X,G18.10)  
13 IPC=0  
IF (IEND.GE.3) GO TO 403  
DO 9 I=1,IC  
9 IF(DABS(WM(I,IC1)).LE.PREC )IPC=IPC+1  
IF(IPC.EQ.IC.OR.IT1.EQ.IT)GO TO 23  
IT1=IT1+1  
C COMPUTE SECOND ORDER PARTIALS  
LLI=1  
LUI=0  
DO 15 I=1,IC  
LLJ=LLI  
LUJ=LUI  
LLI=LLI+MI(I)  
LUI=LUI+MI(I+1)  
DO 16 J=I,IC  
LLJ=LLJ+MI(J)  
LUJ=LUJ+MI(J+1)  
TEMP=0.  
WM(I,J)=0.  
I2=1  
DO 17 J1=LLJ,LUI  
J2=1  
DO 18 J1=LLJ,LUJ  
WM(I,J)=WM(I,J)+W(I1,J1)\*W(I1,J1)  
TEMP=TEMP+WM2(I,I2)\*W(I1,J1)\*WM2(J,J2).  
18 J2=J2+1  
17 I2=I2+1  
TEMP=TEMP/SIGMA  
WM(I,J)=(WM(I,J)/2.0D0)-TEMP  
16 WM(I,J)=2.0D0\*DR(I)\*DR(J)\*WM(I,J)  
WM(I,I)=WM(I,I)-WM(I,IC1)/DR(I)  
DO 15 J=1,I  
WM(I,J)=WM(J,I)  
C SOLVE NEWTON-RAPHSON EQUATIONS FOR DR  
C (ZERO ORDER APPROXIMATION INCLUDED)  
DO 60 I=1,IC  
WM2(I,1)=WM(I,IC1)

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60  WM2(I,2)=WM(I,1)
    DO 61 I=1,IC
    IF(DR(I).LT.EPS)GO TO 61
    CALL INV2(WM,IC,1,I,I,DD,MAXMK1,0)
61  CONTINUE
    DO 62 I=1,IC
    IF(DR(I).LT.EPS)GO TO 63
    DR(I)=DABS(DR(I)+WM(I,IC1))
    GO TO 62
63  DR(I)=DABS(WM2(I,1)/WM2(I,2))
    IF(DR(I).GT.EPS)GO TO 62
    DR(I)=0.
62  D(I)=DR(I)*DR(I)
C    REMOVE ZERO COMPONENTS FROM MODEL
55  DO 54 LOOP=1,IC
    IF(DR(LLOOP).GT.0.)GO TO 54
    IF(IC.EQ.1)GO TO 56
    IEPS=LOOP
205  WRITE(6,205)IEPS
    FORMAT('0COMPONENT ',I4,' IS ZERO AND WILL BE DELETED')
    IX=MI(IEPS+1)
    I1=0
    DO 50 J=1,IEPS
50  I1=I1+MI(J)
    I2=I1+IX
    L=K3-I2
    DO 51 I=1,L
    DO 51 J=1,L
51  R(I1+I,I1+J)=R(I2+I,I2+J)
    IF(IEPS.EQ.1)GO TO 58
    DO 57 I=1,I1
    DO 57 J=1,L
    R(I,I1+J)=R(I,I2+J)
57  R(I1+J,I)=R(I2+J,I)
58  ICT=IC-1
    M=M-IX
    K3=K3-IX
    IF(IEPS.EQ.IC)GO TO 53
    DO 52 I=IEPS,ICT
    DR(I)=DR(I+1)
    D(I)=D(I+1)
52  MI(I+1)=MI(I+2)
    GO TO 53
54  CONTINUE
    GO TO 30
53  IC=IC-1
    IC1=IC1-1
    GO TO 55
56  WRITE(6,206)
206  FORMAT('0ALL VARIANCE COMPONENTS ARE ZERO')
    GO TO 22
C    COMPUTE DELTA H AND DET(H)
30  DO 27 I=1,M
    DO 27 J=I,M
27  WM(I,J)=R(I,J)
    LL=1
    LU=0
    DO 28 I=1,IC
    LL=LL+MI(I)
    LU=LU+MI(I+1)
    DO 28 J=LL,LU
28  WM(J,J)=WM(J,I)+1.D0/D(I)
    CALL CDI(M,WM,TOL,DET,MAXMK1)
    DO 70 I=1,M
    DO 70 J=I,M
70  WM(J,I)=WM(I,J)

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DO 20 I=1,IC
20  DET=DET+DLOG(D(I)**MI(I+1))
C    COMPUTE W + DELTA W
K2=M+1
DO 71 J=1,K2
DO 72 I=1,M
TEMP=0.
DO 73 L=1,M
73  TEMP=TEMP+WM(I,L)*R(L,J)
72  WM2(1,I)=TEMP
DO 71 I=J,K2
TEMP=0
DO 74 L=1,M
74  TEMP=TEMP+R(L,I)*WM2(1,L)
W(J,I)=R(J,I)-TEMP
71  W(I,J)=W(J,I)
GO TO 6
23  CONTINUE
C    COMPUTE LARGE SAMPLE VARIANCE-COVARIANCE MATRIX
DO 93 I=1,IC1
DO 93 J=1,IC1
93  WM(I,J)=0.000
LL=1
LU=0
DO 94 IN=1,IC
LL=LL+MI(IN)
LU=LU+MI(IN+1)
DO 86 I=LL,LU
86  WM(IN,IC1)=WM(IN,IC1)+W(I,I)
LLL=LL
LLU=LU
DO 94 JN=IN,IC
DO 87 I=LL,LU
DO 87 II=LLL,LLU
87  WM(IN,JN)=WM(IN,JN)+W(I,II)*W(II,I)
IF (JN.EQ.IC) GO TO 94
LLL=LLL+MI(JN+1)
LLU=LLU+MI(JN+2)
94  CONTINUE
S1=2*SIGMA
S2=S1*SIGMA
DO 88 I=1,IC
DO 89 J=1,IC
89  WM(I,J)=WM(I,J)/2
88  WM(I,IC1)=WM(I,IC1)/S1
WM(IC1,IC1)=Nk/S2
DO 92 I=1,IC1
DO 92 J=1,IC1
W(I,J)=WM(I,J)
W(J,I)=W(I,J)
92  WM(J,I)=WM(I,J)
CALL INV2(WM,IC1,0,1,IC1,DD,MAXMK1,0)
WRITE(6,215)
215  FORMAT('-',4X,'ESTIMATES',10X,'LARGE SAMPLE VARIANCE-COVARIANCE MA
&TRIX')
DO 85 I=1,IC
85  WRITE(6,207) D(I), (WM(I,J),J=1,IC1)
WRITE(6,207) SIGMA, (WM(IC1,J),J=1,IC1)
C    COMPUTE LARGE SAMPLE VARIANCE-COVARIANCE MATRIX
C    TRANSFORMED TO SIGMA SQUARED (I)
S4=SIGMA*SIGMA
DO 101 I=1,IC
DR(I)=-D(I)/SIGMA
101  WM(I,1)=0.000
DO 102 I=1,IC
DO 102 J=1,IC

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102  WM(I,1)=WM(I,1)+W(I,J)*DR(J)
      T=W(IC1,IC1)
      DO 105 I=1,IC
      T=T+DR(I)*(WM(I,1)+2*W(I,IC1))
      W(I,IC1)=(W(I,IC1)+WM(I,1))/SIGMA
      DO 105 J=1,IC
105  W(I,J)=W(I,J)/S4
      W(IC1,IC1)=T
      DO 104 I=1,IC1
      DO 104 J=1,IC1
104  W(J,I)=W(I,J)
      CALL INV2(W,IC1,0,1,IC1,DD,K4,0)
      WRITE(6,216)
216  FORMAT('-.', 'VARIANCE COMPONENTS', 5X, 'LARGE SAMPLE VARIANCE-COVARIA
      8NCE MATRIX')
      DO 98 I=1,IC
      SIGI=D(I)*SIGMA
98  WRITE(6,207) SIGI, (W(I,J),J=1,IC1)
      WRITE(6,207) SIGMA, (W(IC1,J),J=1,IC1)
C    TRANSFORM R ( M1 X M1 ) BACK TO ITS ORIGINAL FORM
      M1=M+1
      M2=M+2
      MMIN=M-1
      DO 65 I=1,M
      DO 65 J=1,K
      W(I,J)=0.0D0
      DO 65 II=1,K
65  W(I,J)=W(I,J)+P(I,M1+II)*R(M1+II,M1+J)
      DO 66 I=1,M
      DO 66 J=1,M
      DO 66 II=1,K
66  R(I,J)=R(I,J)+W(I,II)*R(M1+II,J)
      DO 25 I=1,M
      DO 67 II=1,K
67  R(I,M1)=R(I,M1)+W(I,II)*R(M1+II,M1)
25  R(M1,I)=R(I,M1)
      R(M1,M1)=R(M1,M1)+YY
      CALL INV2(R,K,0,M2,K3,DD,K4,M1)
C    COMPUTE FIXED EFFECTS
      LL=1
      LU=0
      DO 36 I=1,IC
      LL=LL+MI(I)
      LU=LU+MI(I+1)
      DO 36 J=LL,LU
36  R(J,J)=R(J,J)+1.0D0/D(I)
      CALL CDI(M,R,TOL,DET,K4)
      DO 90 I=1,M
      DO 90 J=1,M
90  R(J,I)=R(I,J)
      DO 40 I=1,K
      DO 40 J=1,M
      WM(I,J)=0.0D0
      DO 40 II=1,M
40  WM(I,J)=WM(I,J)+R(M1+I,II)*R(II,J)
      DO 41 I=1,K
      DO 41 J=1,K
      W(I,J)=0.0D0
      DO 41 II=1,M
41  W(I,J)=W(I,J)+WM(I,II)*R(II,M1+J)
      DO 49 I=1,K
      W(I,K1)=0.0D0
      DO 42 II=1,M
42  W(I,K1)=W(I,K1)+WM(I,II)*R(II,M1)
49  W(K1,I)=W(I,K1)
      AB=0.0D0

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DO 43 J=1,M
WM2(1,J)=0.0D0
DO 45 II=1,M
45 WM2(1,J)=WM2(1,J)+R(M1,II)*R(II,J)
43 AB=AB+WM2(1,J)*R(J,M1)
W(K1,K1)=AB
DO 46 I=1,K
W(I,K1)=R(M1+I,M1)-W(I,K1)
W(K1,I)=W(I,K1)
DO 46 J=1,K
46 W(I,J)=R(M1+I,M1+J)-W(I,J)
W(K1,K1)=R(M1,M1)-W(K1,K1)
GO TO 21
C      COMPUTE FIXED EFFECTS WHEN ALL VARIANCE COMPONENTS ARE ZERO
22 DO 38 I=1,K
W(I,K1)=R(I+1,1)
W(K1,I)=W(I,K1)
DO 38 J=1,K
38 W(I,J)=R(I+1,J+1)
W(K1,K1)=R(1,1)+YY
CALL INV2(W,K,0,1,K,DD,K4,0)
21 CALL INV2(W,K,1,1,K,DD,K4,0)
C      COMPUTE VARIANCE-COVARIANCE MATRIX OF THE FIXED EFFECTS
DO 19 I=1,K
DO 19 J=1,K
19 W(I,J)=W(I,J)*SIGMA
WRITE(6,212)
212 FORMAT(' ',3X,'FIXED EFFECTS',RX,'VARIANCE-COVARIANCE MATRIX')
DO 47 I=1,K
47 WRITE(6,207) W(I,K1),(W(I,J),J=1,K)
207 FORMAT('0',G17.8,3X,6G17.8/' ',20X,6G17.8)
RETURN
403 WRITE(6,208)
208 FORMAT(' ', 'THE LIKELIHOOD IS BEING MINIMIZED, A POSSIBLE CORRECTI
80N IS TO TRY DIFFERENT INITIAL VALUES FOR THE GAMMA VECTOR')
RETURN
END
SUBROUTINE INV2(A,N,M,N1,N2,D,M1,NN)

IMPLICIT REAL*8(A-H,O-Z)
DIMENSION A(M1,M1)
N3=1
IF (NN.NE.0) N3=N1
L=NN+N+M
LN=NN+N
D=1.0D0
DO 1 I=N1,N2
D=D*A(I,I)
A(I,I)=1.0D0/A(I,I)
DO 2 J=N3,L
IF (J-I)3,2,3
3 A(I,J)=A(I,J)*A(I,I)
2 CONTINUE
DO 1 J=N3,LN
IF (J-I)4,1,4
4 DO 6 K=N3,L
IF (K-I)5,6,5
5 A(J,K)=A(J,K)-A(J,I)*A(I,K)
6 CONTINUE
A(J,I)=-A(J,I)*A(I,I)
1 CONTINUE
RETURN
END

SUBROUTINE CDI(N,A,TOL,D,MAX)
C      INVERSION BY CHOLESKY DECOMPOSITION

```

```

      IMPLICIT REAL*4(A-H,O-Z)
      DIMENSION A(MAX,MAX)
      C      TRIANGULAR DECOMPOSITION
      DO 2 K=1,N
      K1=K+1
      IF(A(K,K).LE.TOL)GO TO 5
      A(K,K)=DSORT(A(K,K))
      IF(K.EQ.N)GO TO 3
      DO 1 J=K1,N
      1  A(K,J)=A(K,J)/A(K,K)
      DO 2 I=K1,N
      DO 2 J=I,N
      2  A(I,J)=A(I,J)-A(K,I)*A(K,J)
      C      COMPUTE DETERMINANT OF A
      3  D=0.000
      DO 4 K=1,N
      4  D=D+DLOG(A(K,K))
      D=D+D
      C      BACKSOLUTION FOR R INVERSE
      DO 6 K=1,N
      L=N+1-K
      A(L,L)=1.00/A(L,L)
      IF(L.EQ.1)GO TO 8
      DO 6 I=2,L
      I2=L+1-I
      T=0.
      DO 7 M=2,I
      J=L-M+2
      7  T=T-A(I2,J)*A(J,L)
      6  A(I2,L)=T/A(I2,I2)
      8  CONTINUE
      C      MULTIPLICATION FOR A INVERSE
      DO 9 I=1,N
      DO 9 J=I,N
      T=0.
      DO 10 K=J,N
      10  T=T+A(I,K)*A(J,K)
      9  A(I,J)=T
      RETURN
      5  WRITE(6,207)
      207 FORMAT(' MATRIX Q IS NOT POSITIVE DEFINITE')
      STOP
      END

```

//LKED.SYSIN DD \*

//GO.SYSIN DD \*

OSTLE'S 14 PINE BOARD DATA

14 1 1 5 20 1 10

5

(6F3.0,F5.1)

(' ',5F3.0,3Y,F3.0,3Y,F5.1)

1 1 7.3

1 1 8.3

1 1 7.6

1 1 8.4

1 1 8.3

1 1 5.4

1 1 7.4

1 1 7.1

1 1 8.1

1 1 6.4

1 1 7.9

1 1 9.5

1 1 10.0

1 1 7.1

1.0

### Credit

Portions of the coding have been taken from a program prepared by W. J. Hemmerle and designed after the Hemmerle and Hartley (1973) algorithm.

### References

- Corbeil, R. R. and S. R. Searle (1974). Restricted maximum likelihood (REML) estimation of variance components in the mixed model. Paper No. BU-538-M in the Biometrics Unit Mimeo Series, Cornell University, Ithaca, New York.
- Hemmerle, W. J. and H. O. Hartley (1973). Computing maximum likelihood estimates for the mixed A.O.V. model using the W-transformation. Technometrics 15:819-832.